

POD-based reduction methods, the Quasi-continuum method and their resemblance

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RUES

RESEARCH UNIT
IN ENGINEERING
SCIENCES



UNIVERSITÉ DU
LUXEMBOURG

Large nonlinear models are inefficient due to:

- 1. Many DOFs**
- 2. Many integration points**

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1. Many DOFs

Interpolation

2. Many integration points

Reduced integration

1. Applications: discrete materials & discrete models

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2. The Quasicontinuum method:

interpolation & integration

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3. POD-based reduction methods:

interpolation & integration

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4. Concluding remarks

1. Applications: discrete materials & discrete models

2. The Quasicontinuum method:

interpolation & integration

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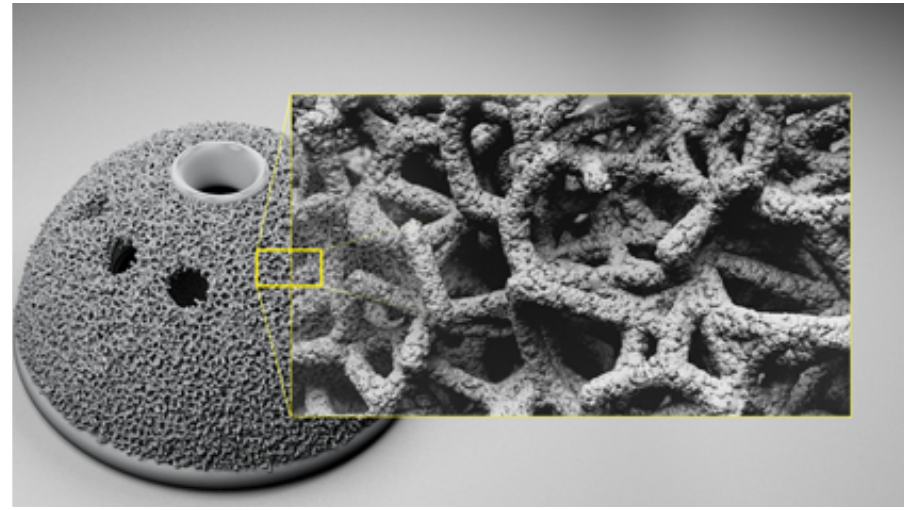
interpolation & integration

4. Concluding remarks

Some discrete materials

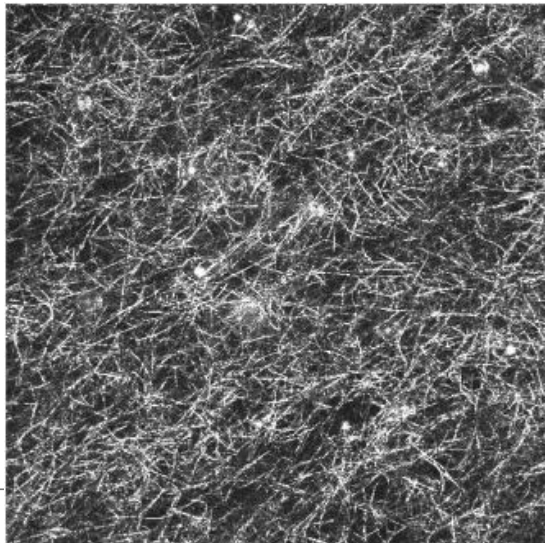


Foams

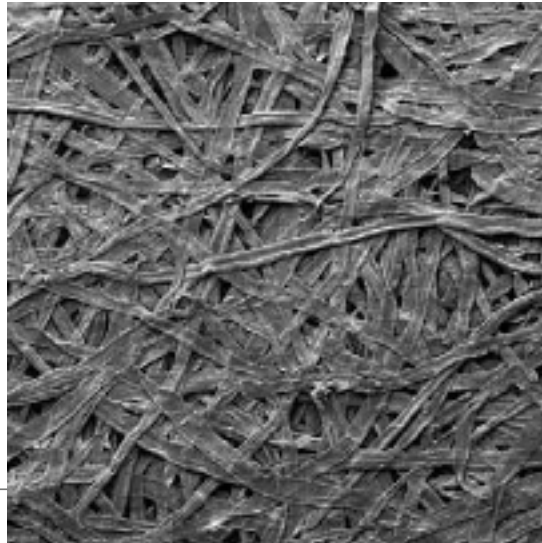


Additive manufacturing

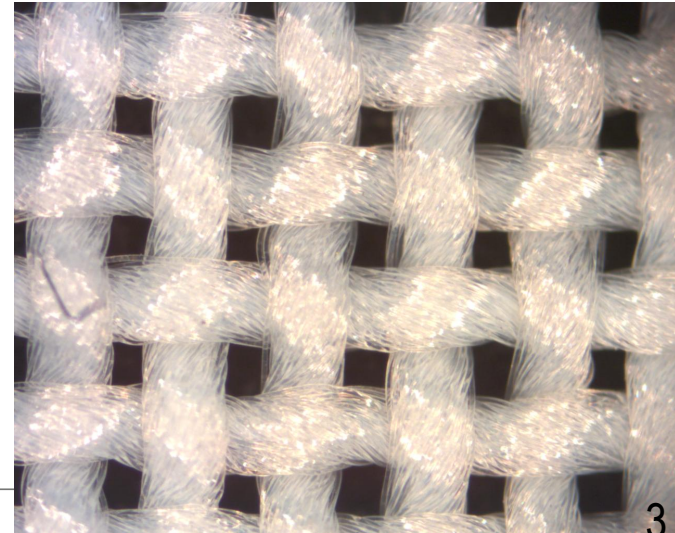
Collagen

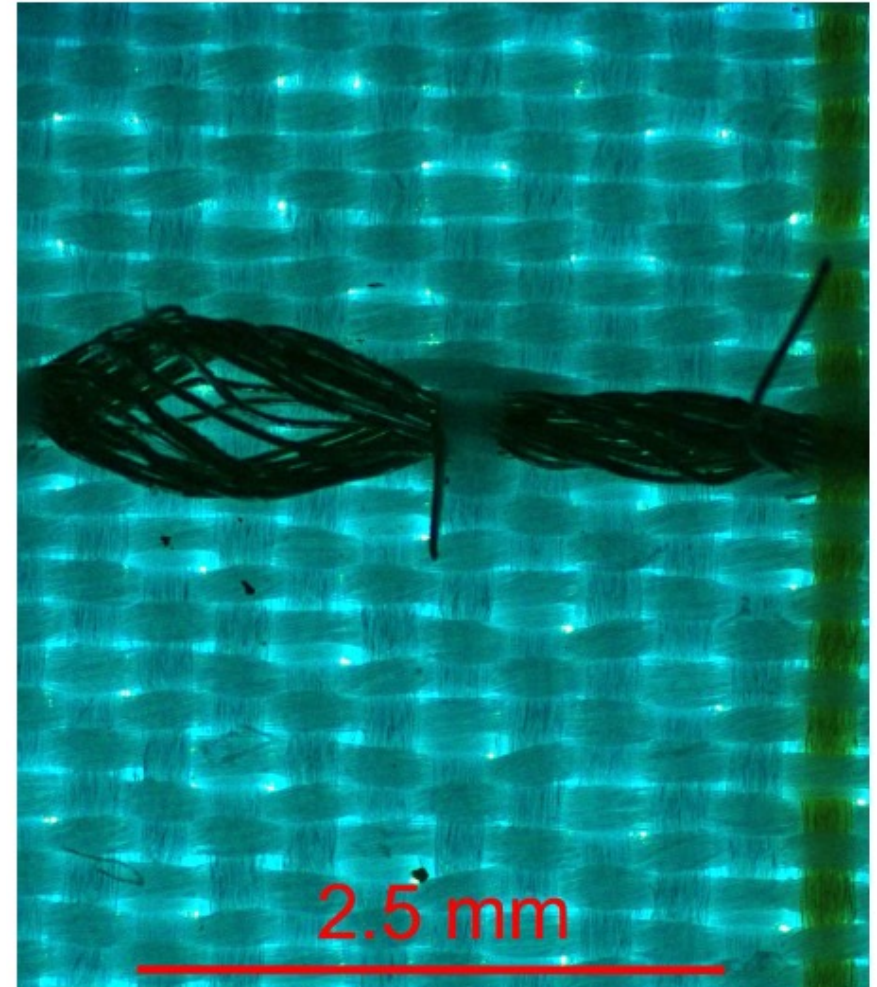
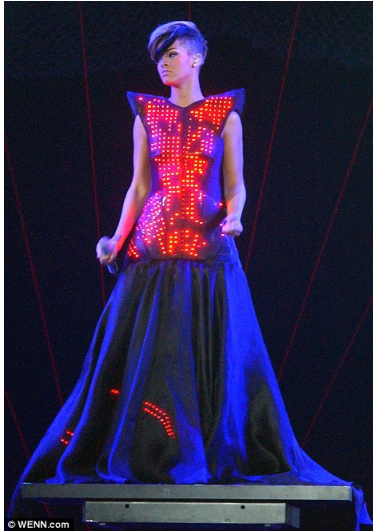


Paper/cardboard

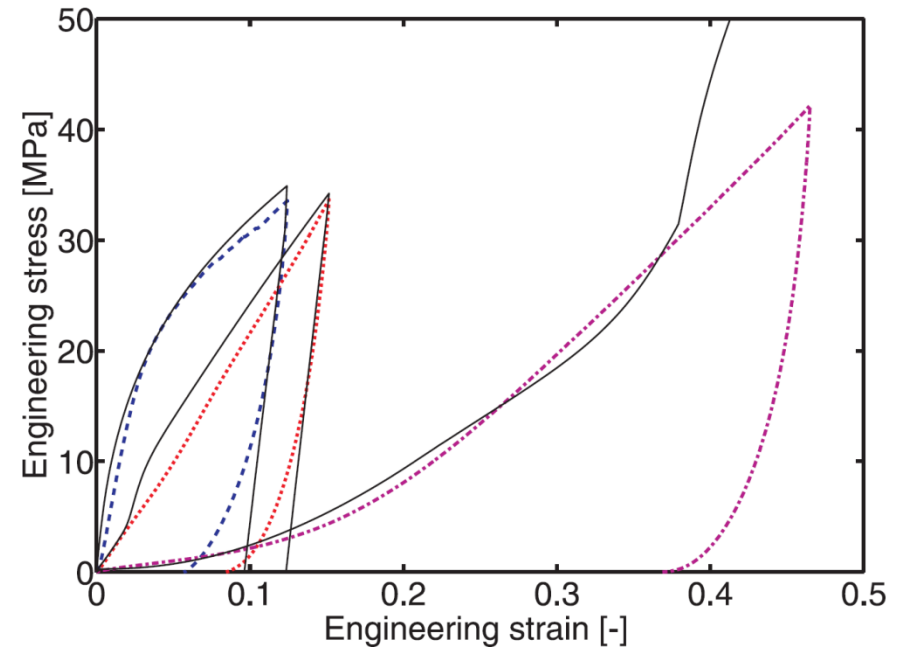
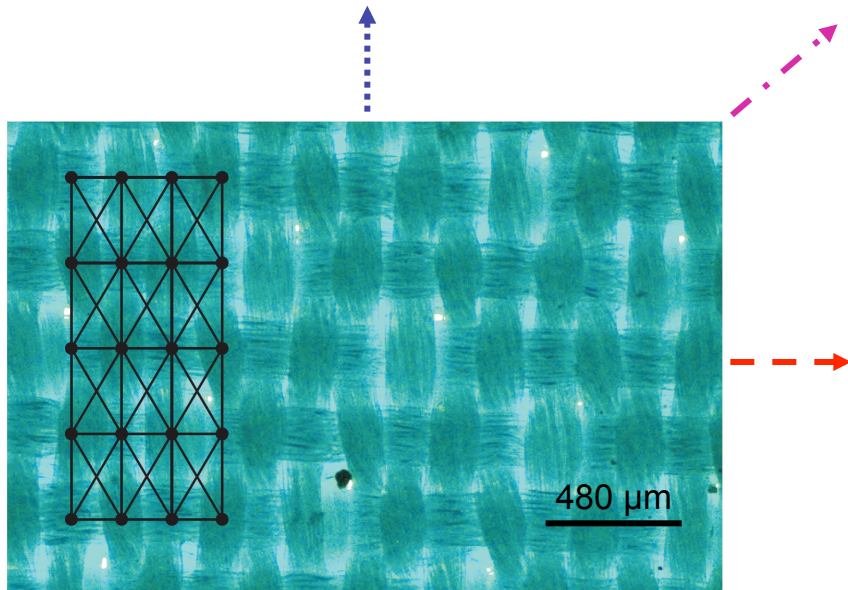


Textiles





Elastoplastic spring lattice



1. Applications: discrete materials & discrete models

2. The Quasicontinuum method:

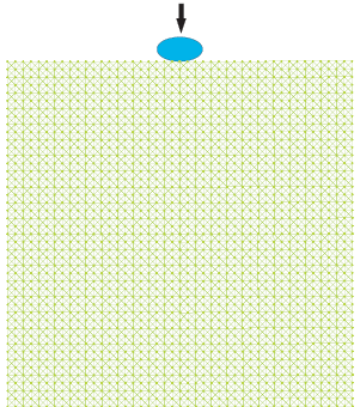
interpolation & integration

3. POD-based reduction methods:

interpolation & integration

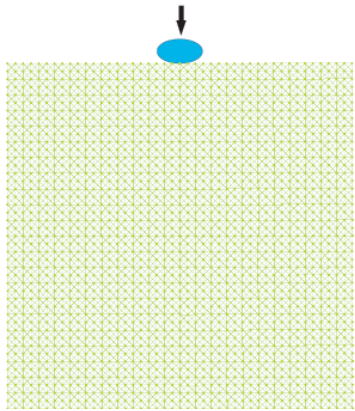
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Direct simulation of elastic lattice



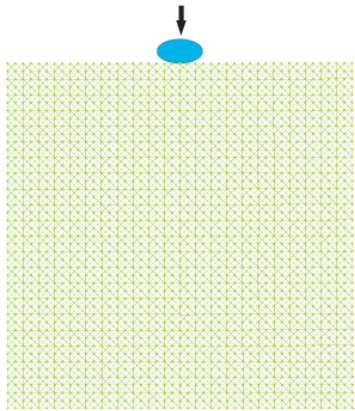
$$E_{tot}(u) = E_{int}(u) - f_{ex}^T u$$

Direct simulation of elastic lattice



$$E_{tot}(u) = E_{int}(u) - f_{ex}^T u$$
$$E_{tot}(u) = \sum_{i=1}^n E_i(u) - f_{ex}^T u$$

Interpolation

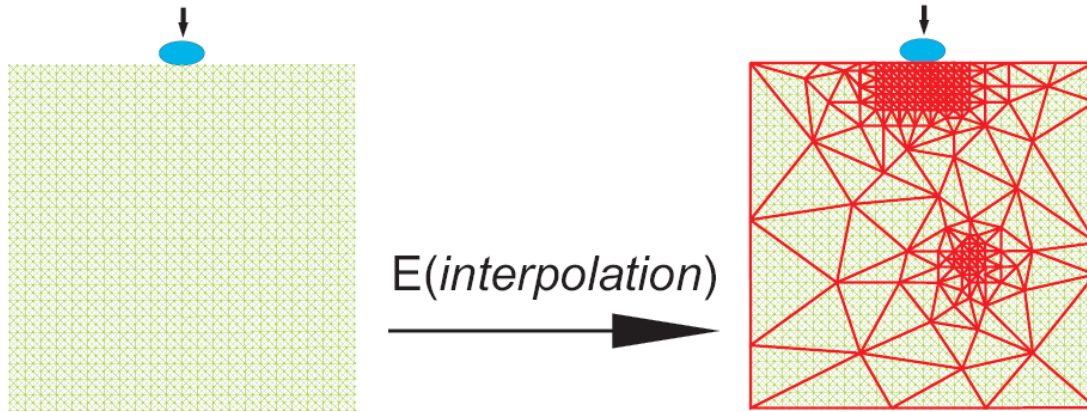


$$E_{tot}(u) = \sum_{i=1}^n E_i(u) - f_{ex}^T u$$

$$E_{tot}(Nu_t) = \sum_{i=1}^n E_i(Nu_t) - f_{ex}^T Nu_t$$

$$u = Nu_t$$

Linear interpolation



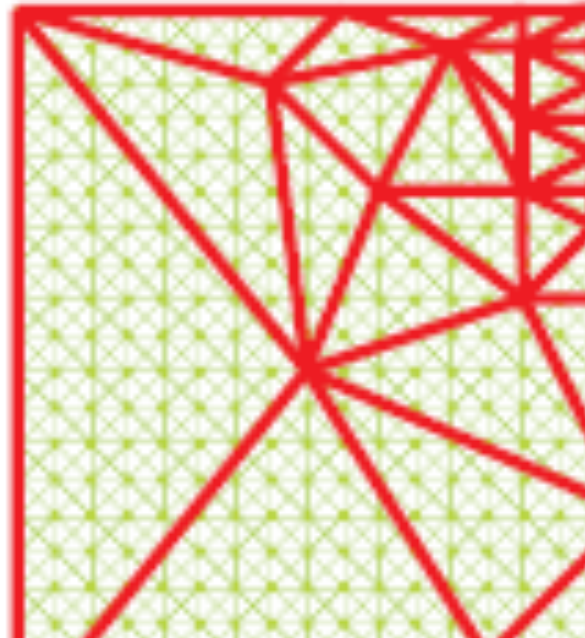
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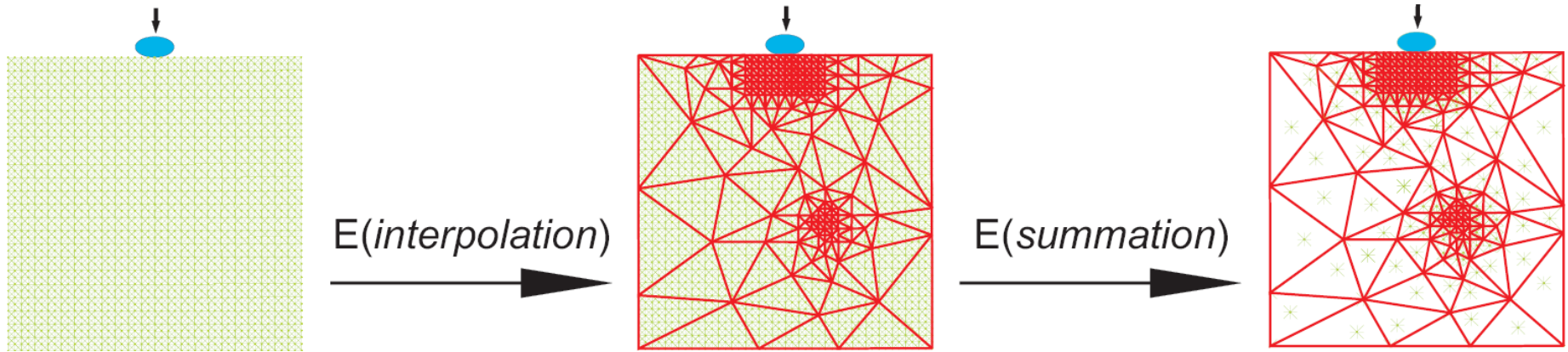
$$u = Nu_t$$

Linear interpolation

$$u = Nu_t$$



Integration for linear triangles

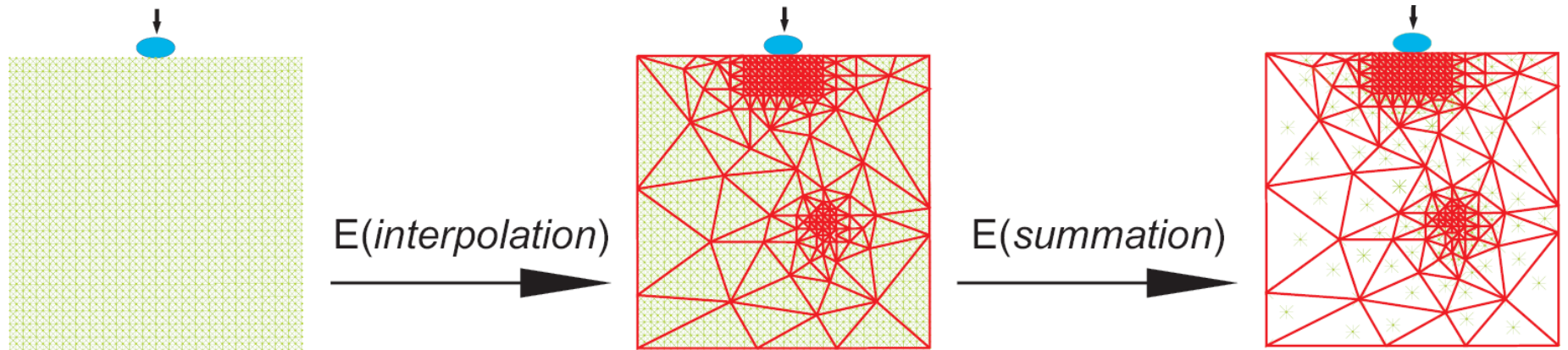


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Integration for linear triangles



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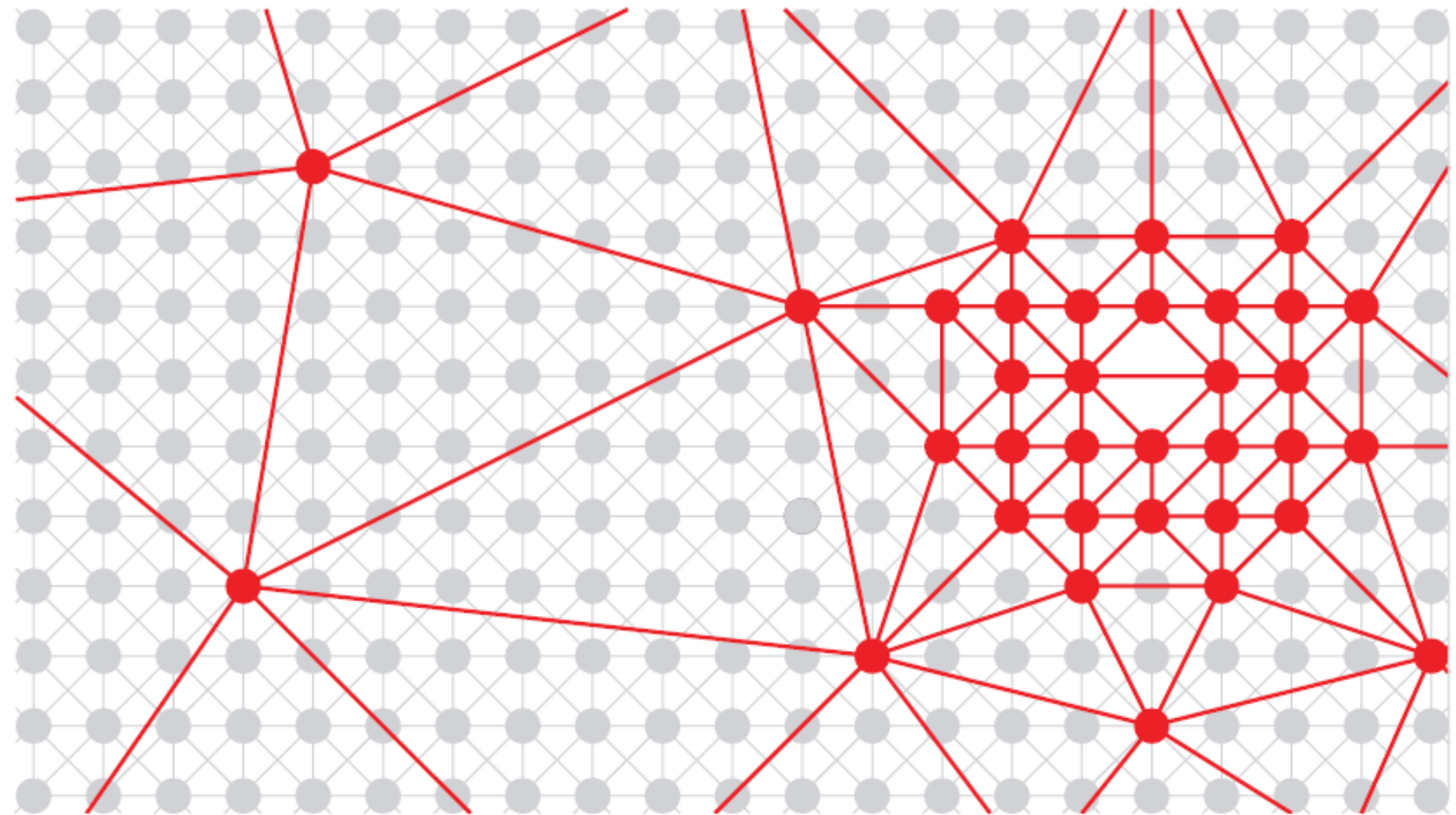
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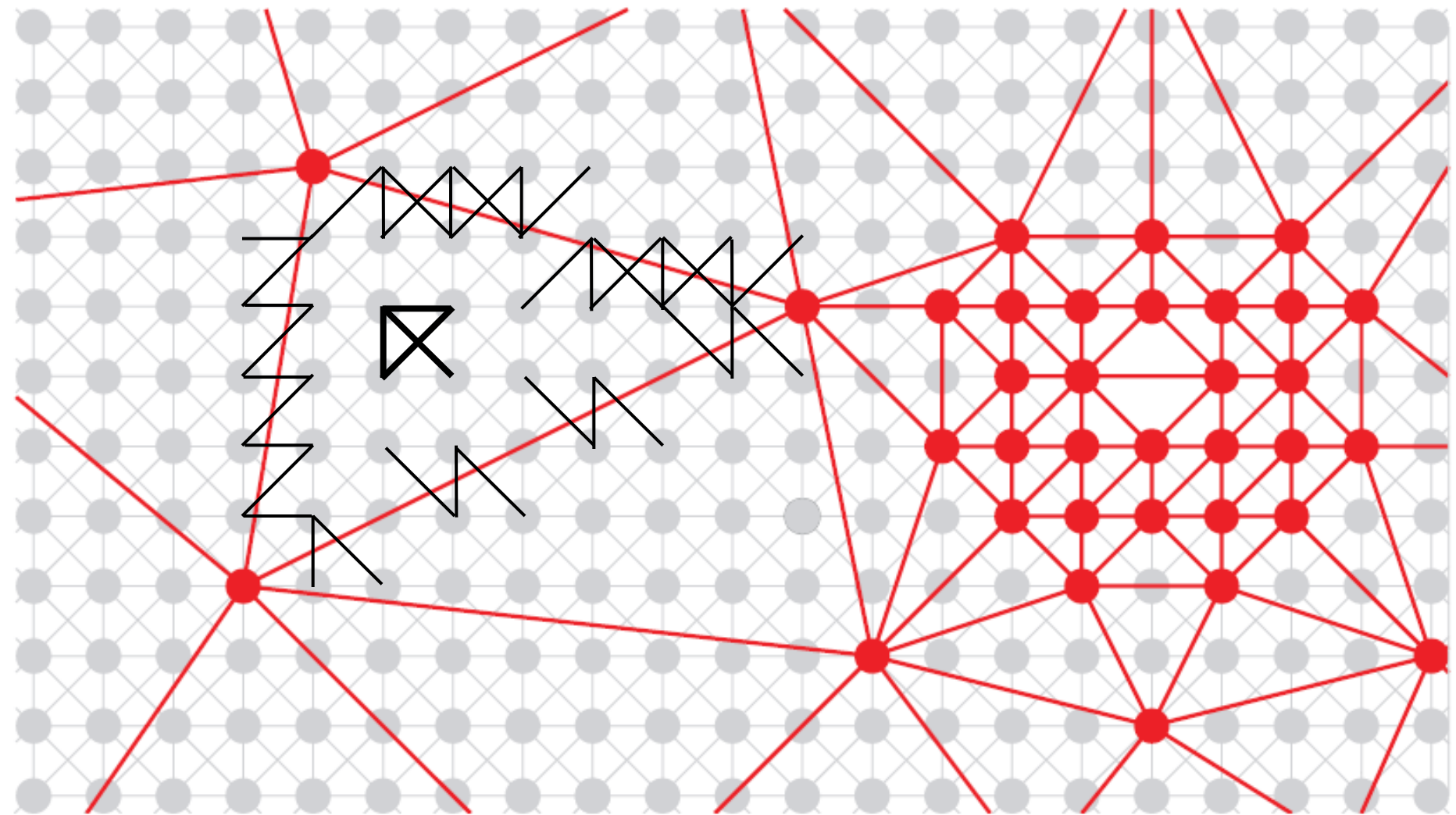
$$u = Nu_t$$

$$\sum_{i=1}^n E_i(Nu_t) \approx \sum_{i=1}^s w_i E_i(Nu_t)$$

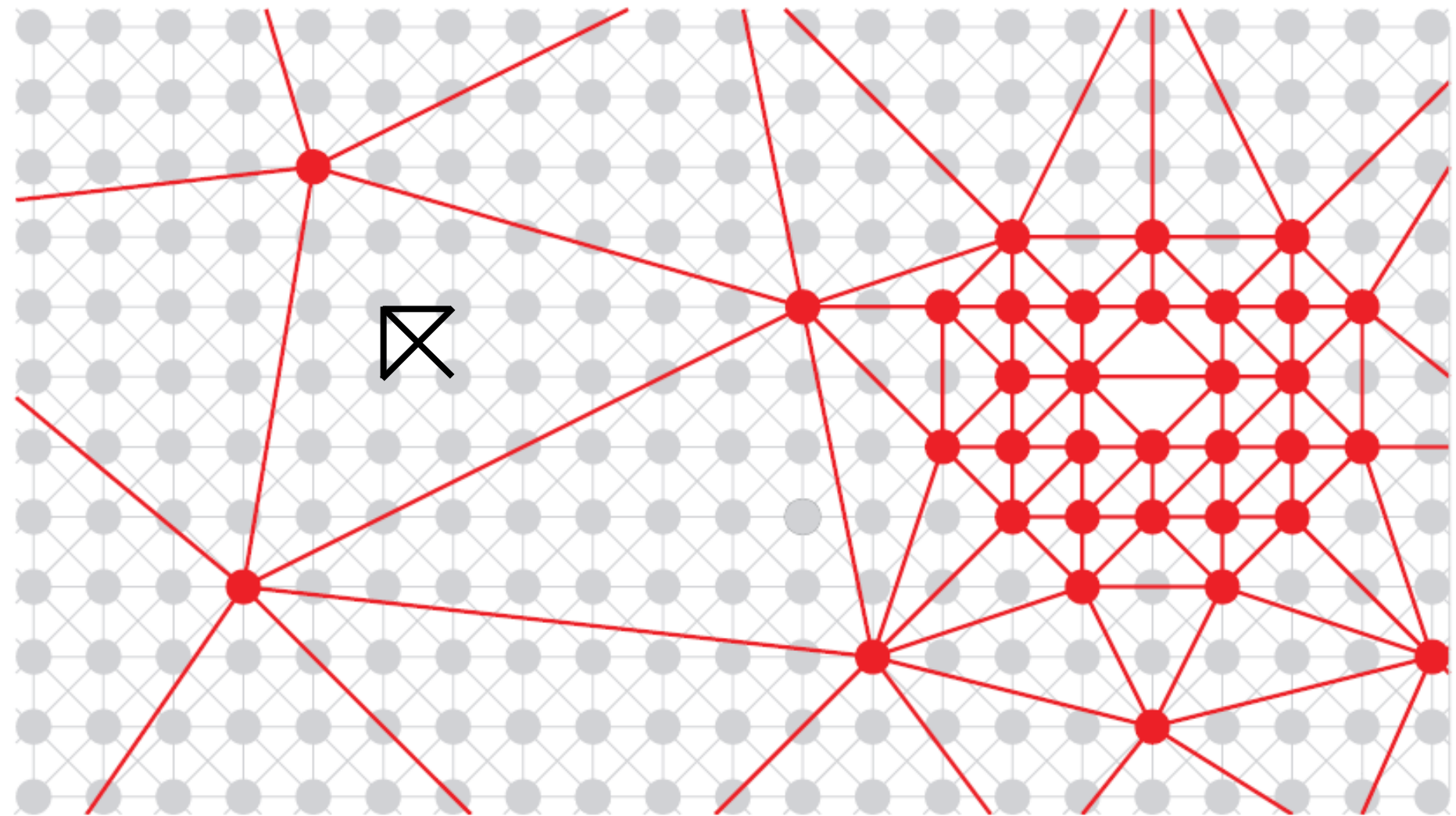
Integration for linear triangles



Integration 1 for linear triangles

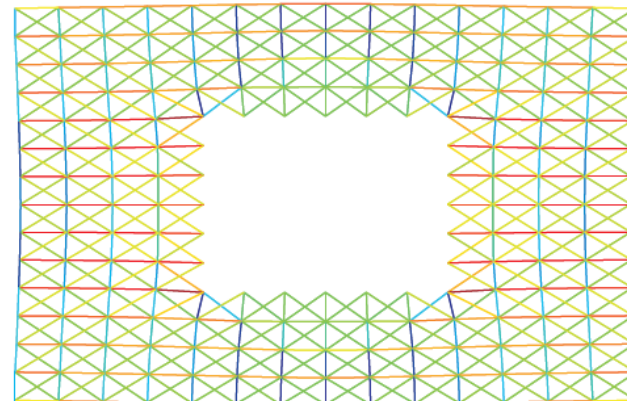
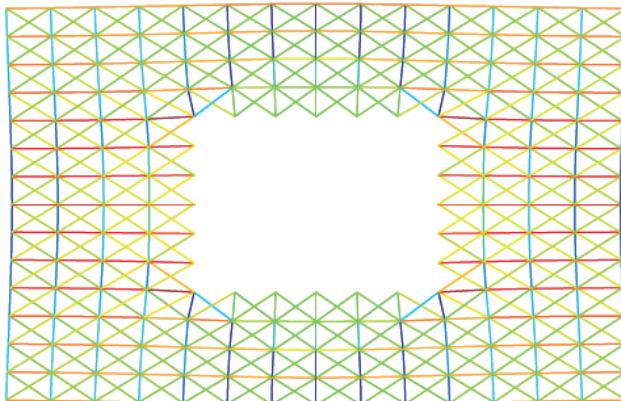
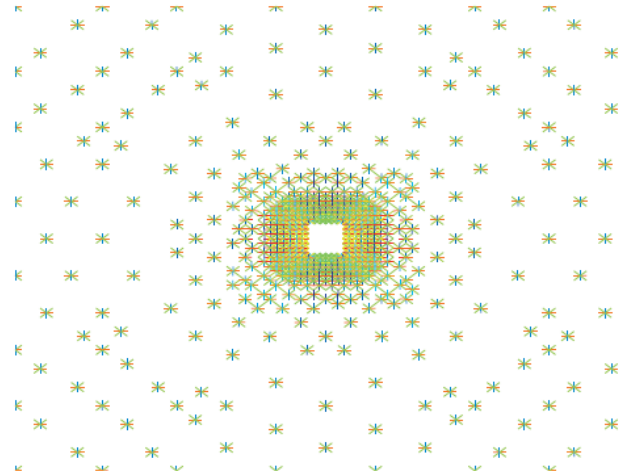
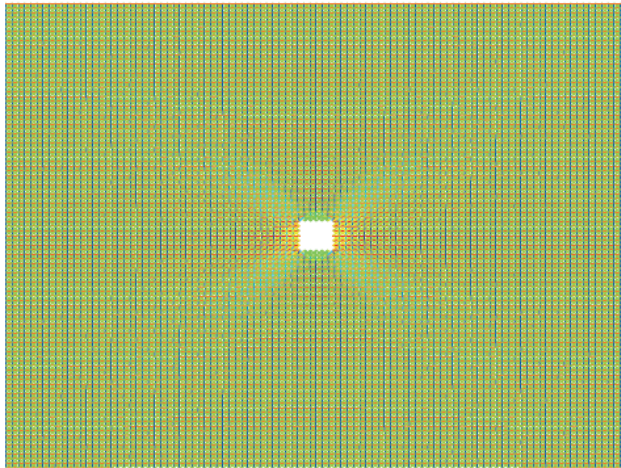
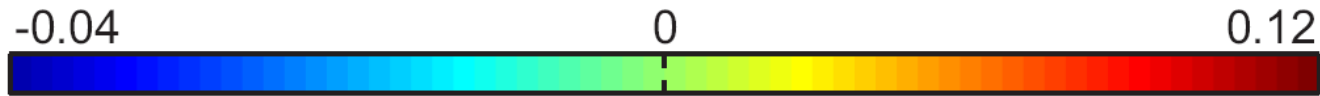


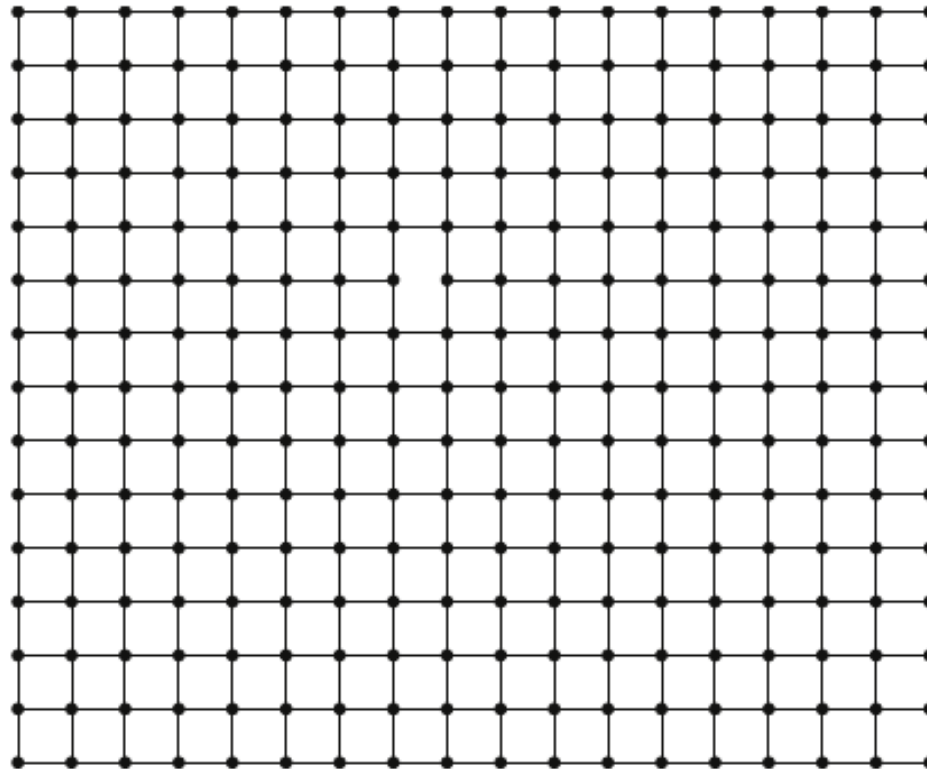
Integration 2 for linear triangles



Accuracy and efficiency

Plastic strain at 10% horizontal stretch

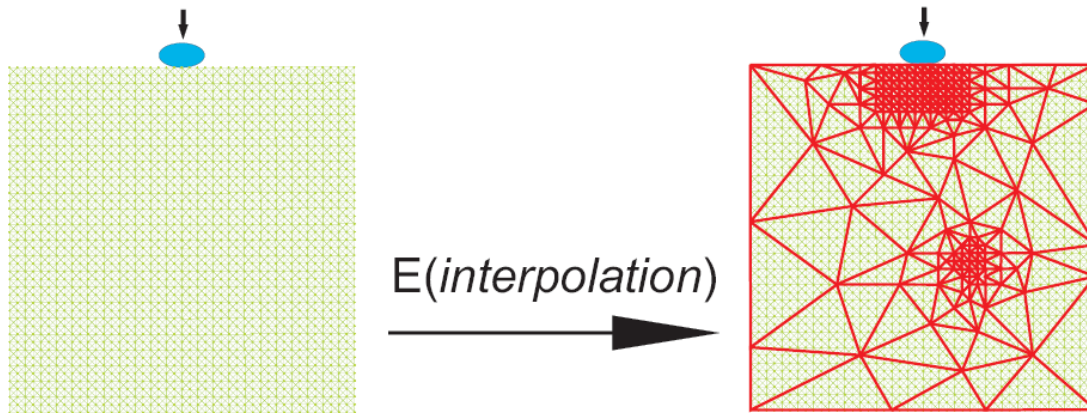




Euler Bernoulli beams:

- Hermite interpolation in each beam
- nodal displacements
- nodal rotations

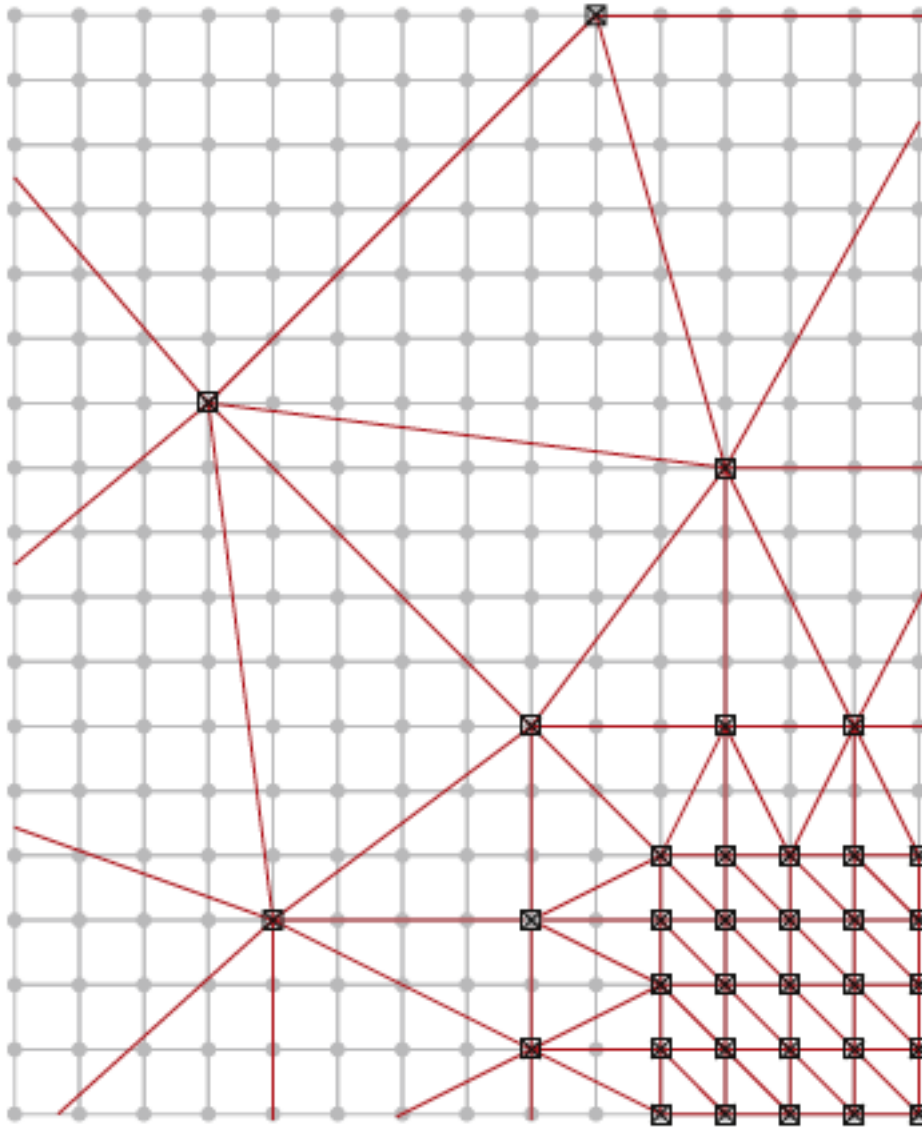
Interpolation



$$E_{tot}(u) = \sum_{i=1}^n E_i(u) - f_{ex}^T u$$

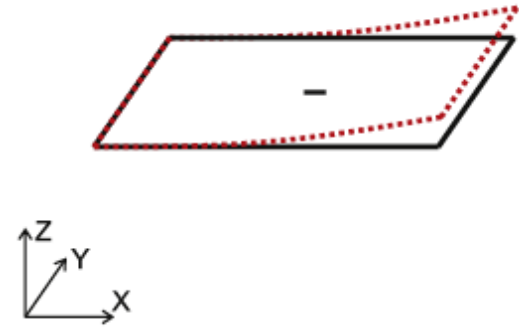
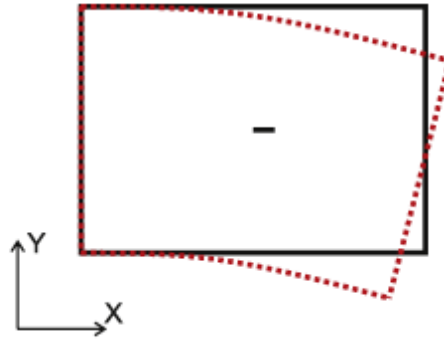
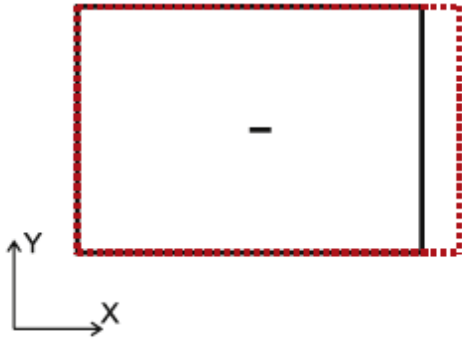
$$E_{tot}(Nu_t) = \sum_{i=1}^n E_i(Nu_t) - f_{ex}^T Nu_t$$

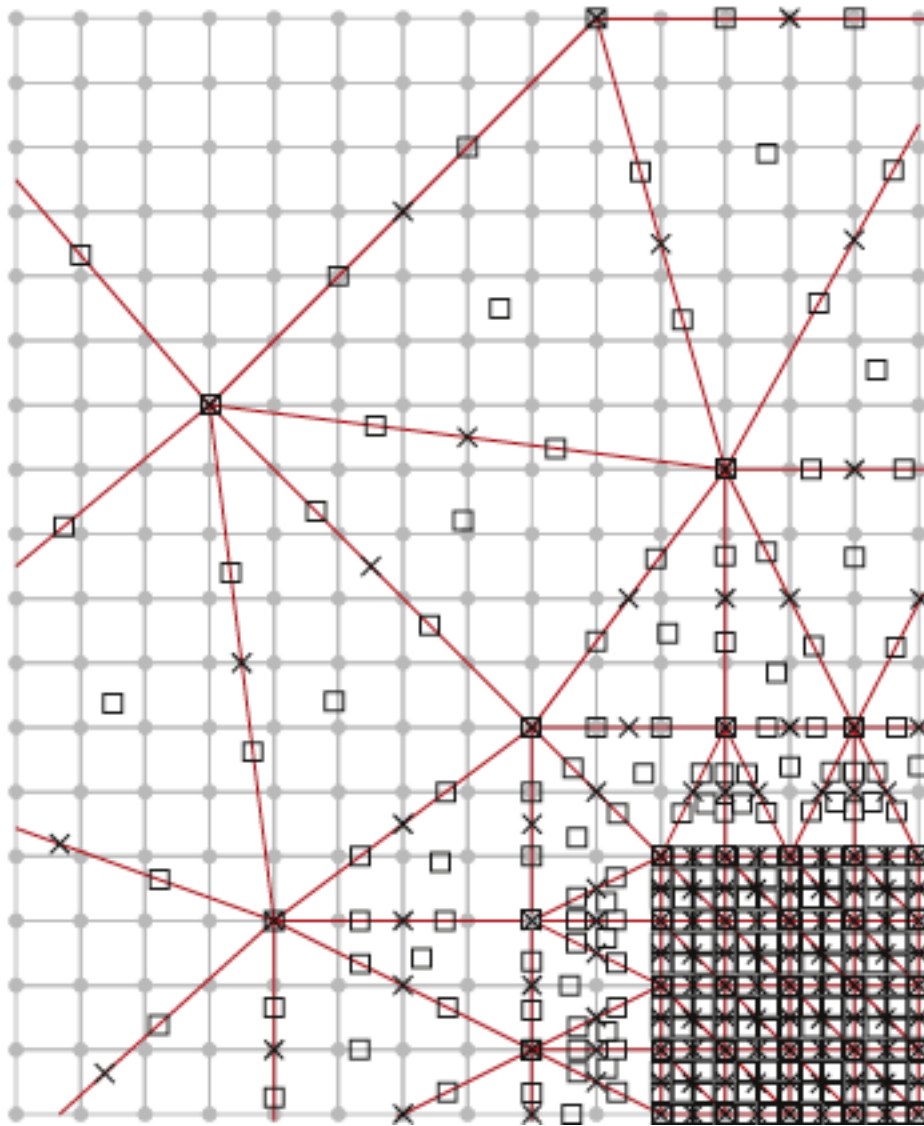
$$u = Nu_t$$



Nodal displacements: Linear
Nodal rotations: Linear
Conforming triangulations

Test cases

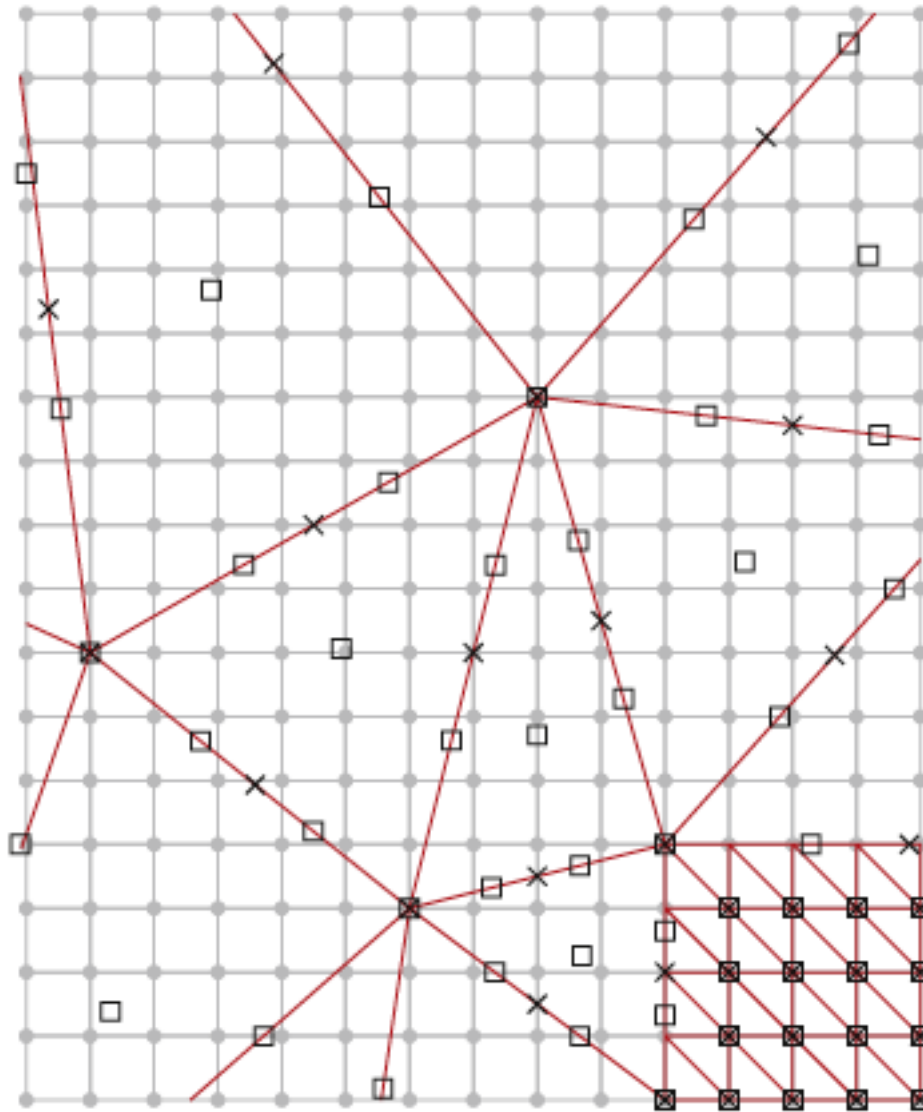




Nodal displacements: Cubic

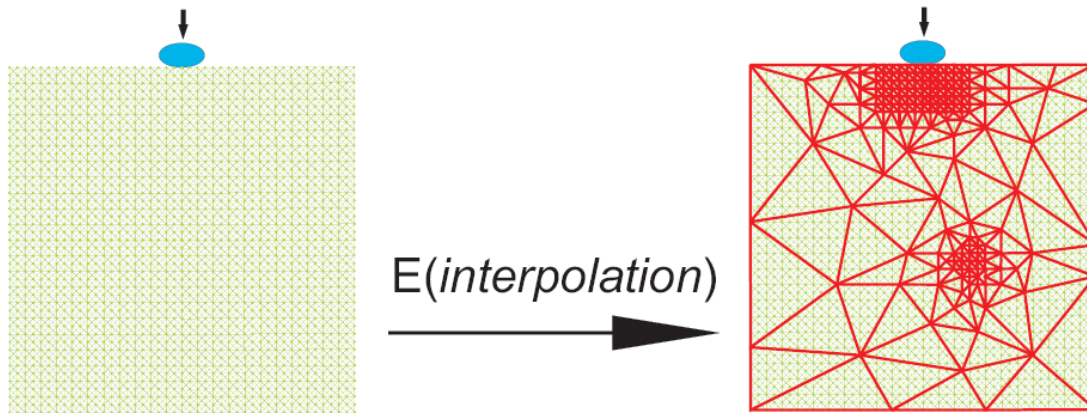
Nodal rotations: Quadratic

Conforming triangulations



Nodal displacements: Cubic
Nodal rotations: Quadratic
Non-conforming triangulations

Interpolation



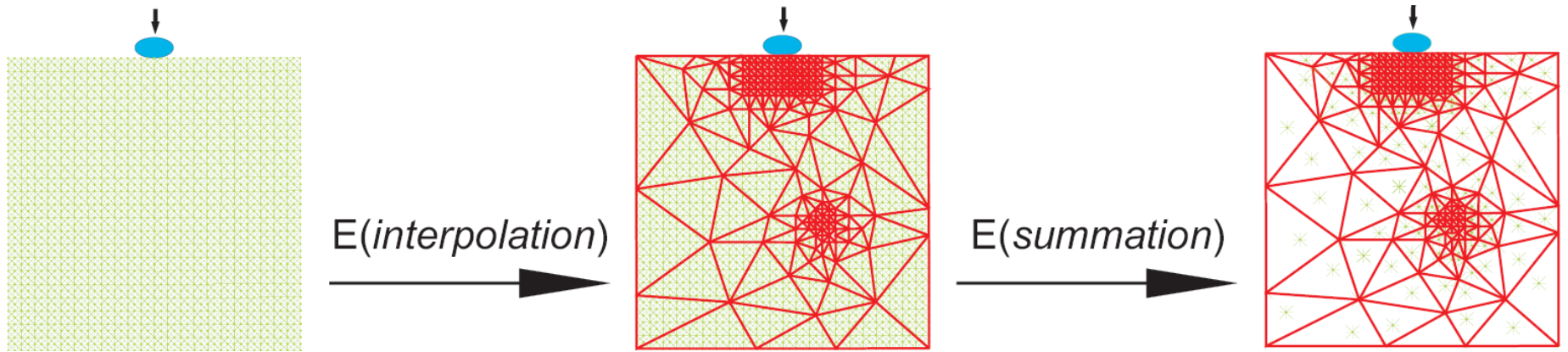
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QC method for planar beam lattice

Integration



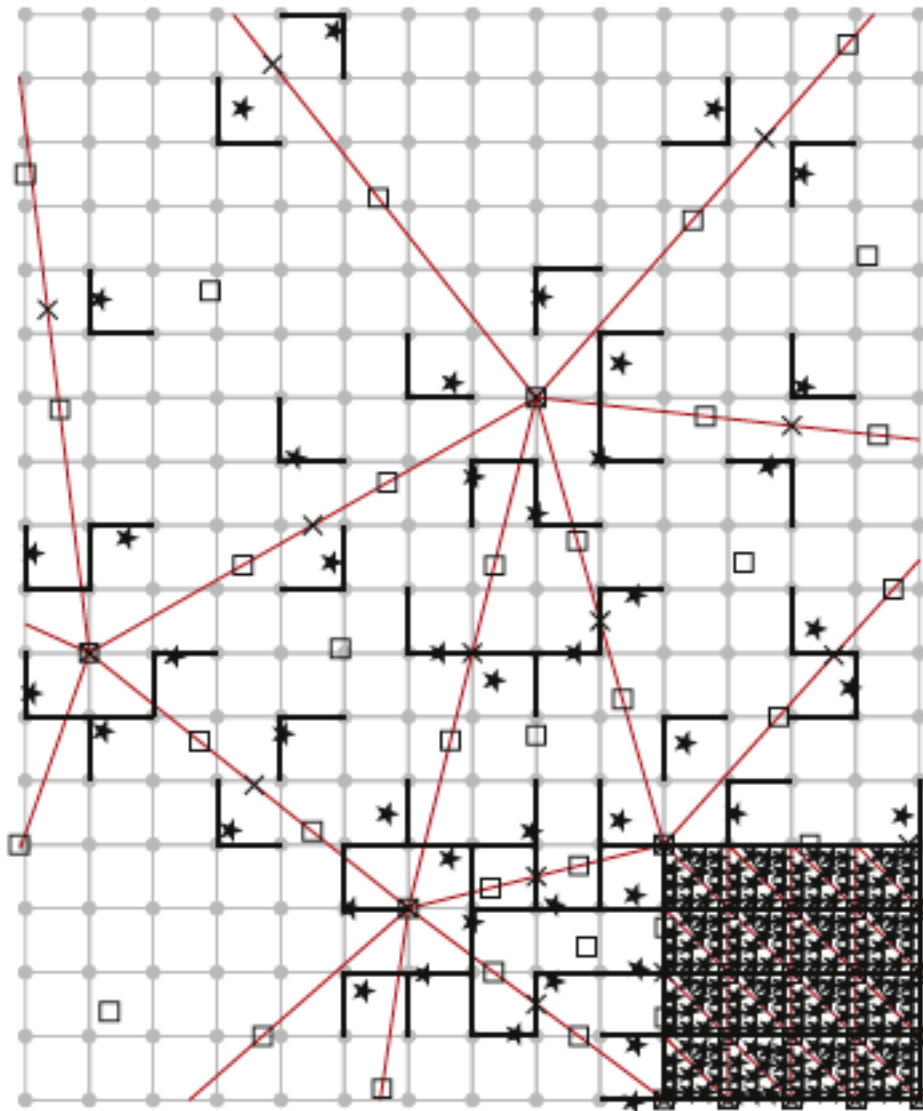
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$$u = Nu_t$$

$$\sum_{i=1}^n E_i(Nu_t) \approx \sum_{i=1}^s w_i E_i(Nu_t)$$



**Sampling beams near
Gauss points**

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E. Schenone, J.S. Hale, S. Bordas

**Largescale
model**

$E(\text{interpolation})$

?

$E(\text{summation})$

?

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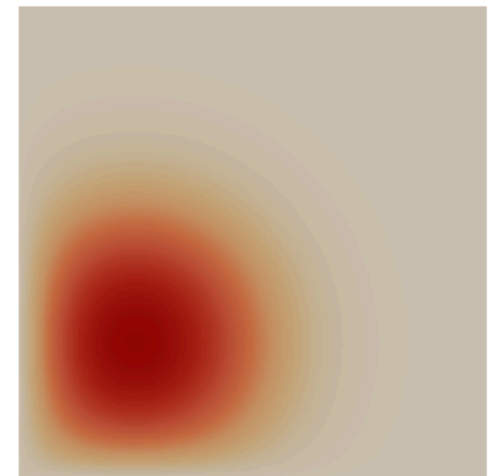
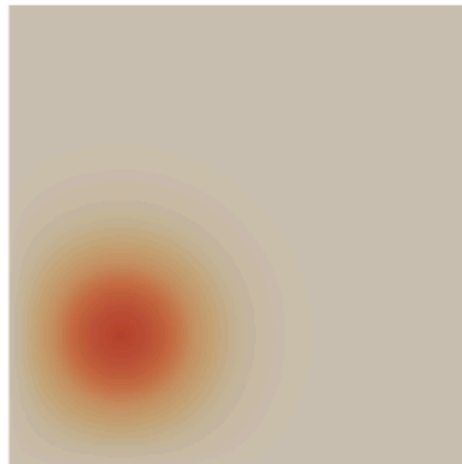
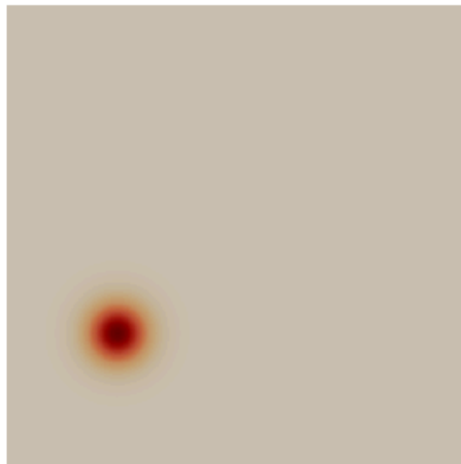
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$$u = Nu_t$$

$$\sum_{i=1}^n E_i(Nu_t) \approx \sum_{i=1}^s w_i E_i(Nu_t)$$

Nonlinear reaction diffusion

$$K\Delta u = f(u)$$



time 

**Largescale
model**

$E(\textit{interpolation})$

?

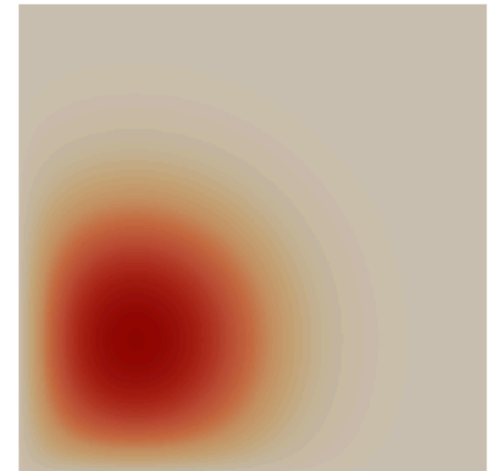
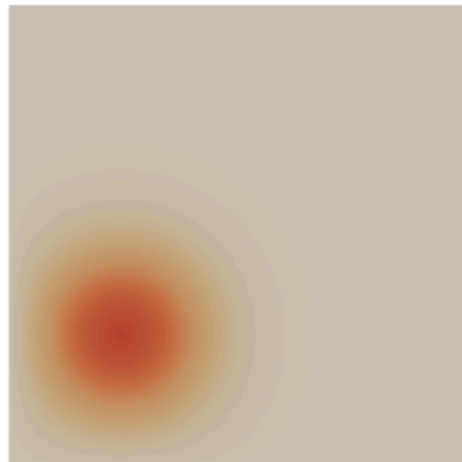
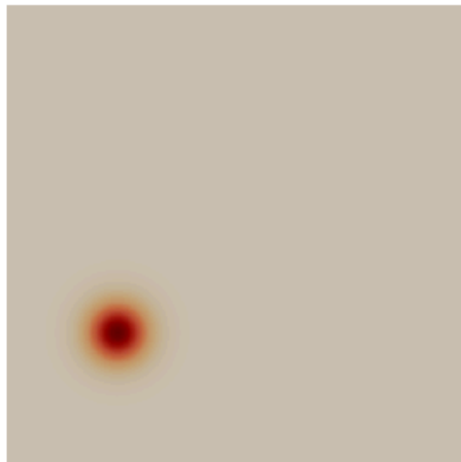
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How to find N: *Offline snapshot/training generation*

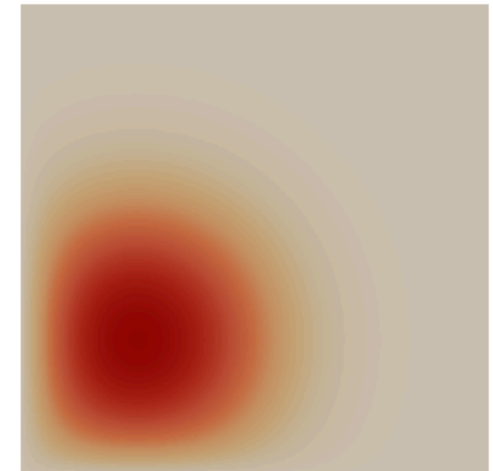
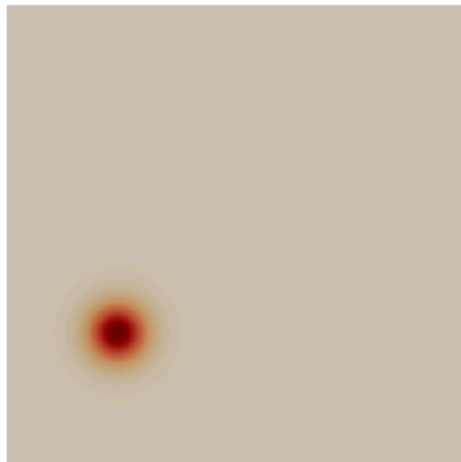
$$K \Delta u = f(u) \quad \xrightarrow{\text{many results: } u_1, u_2, \dots, u_{150}}$$



time 

How to find N: *Offline snapshot/training generation*

$K\Delta u = f(u)$ $\xrightarrow{\text{many results: } u_1, u_2, \dots, u_{150}}$ $N = [u_1, u_2, \dots, u_{150}]$
reduced basis



time \longrightarrow

Or apply first singular value decomposition to $N = [u_1, u_2, \dots, u_{150}]$

and use the modes with the largest eigenvalues $N = [\varphi_1, \varphi_2, \dots, \varphi_{30}]$

Note: N is full

**Largescale
model**

$E(\text{interpolation})$

**Offline training
to find modes**

$E(\text{summation})$

?

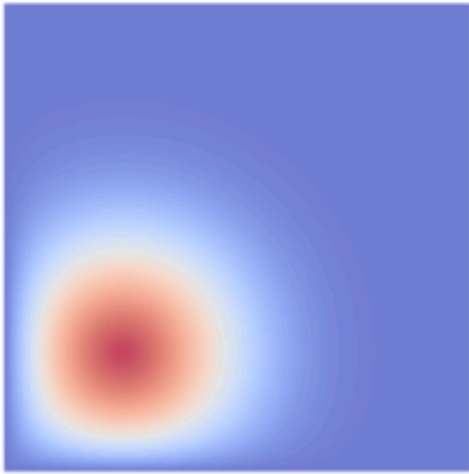
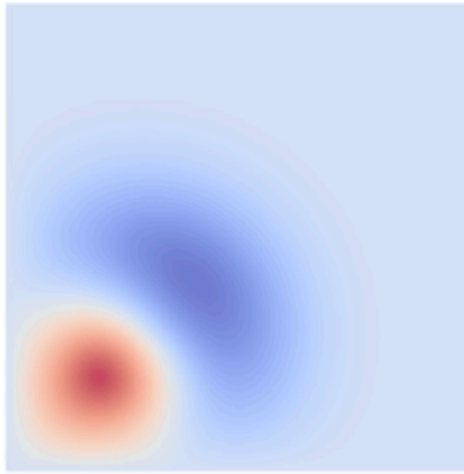
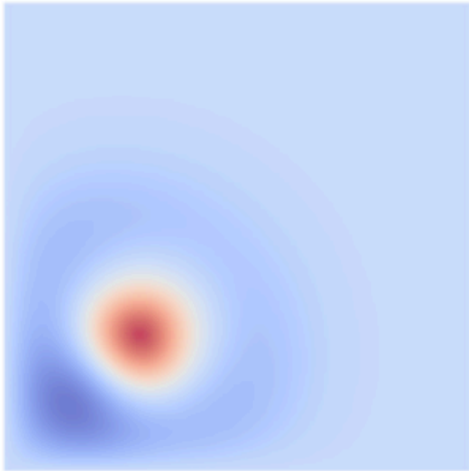
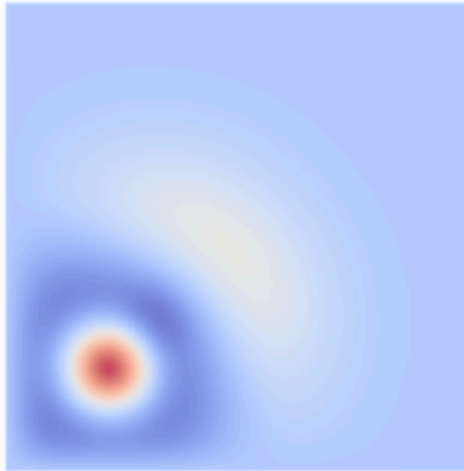
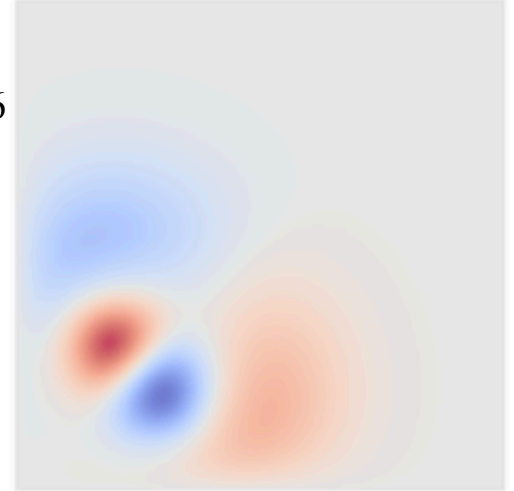
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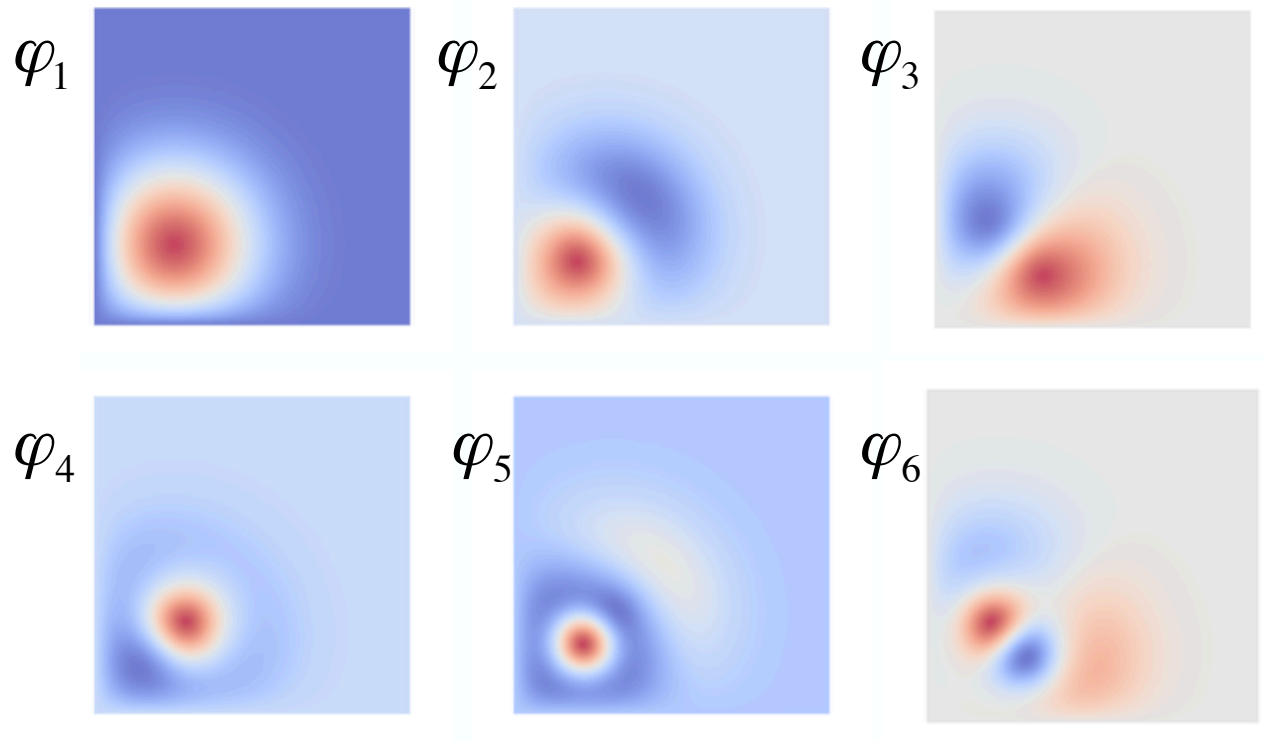
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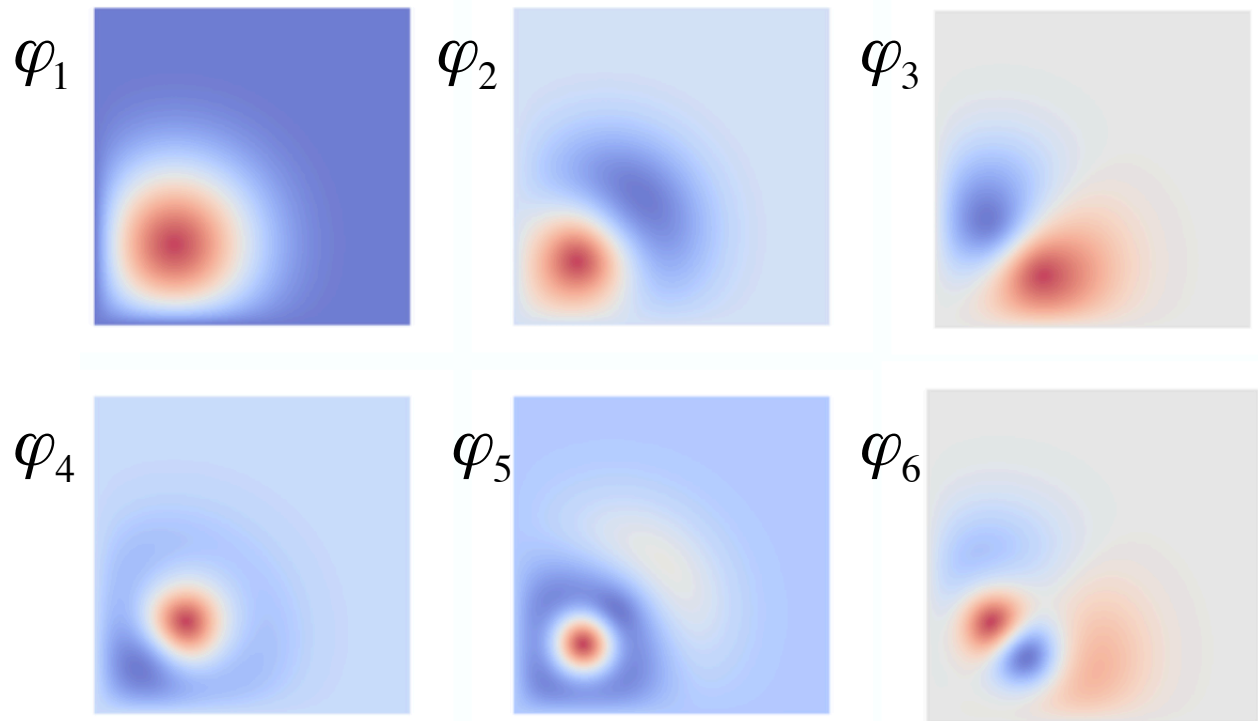
φ_1  φ_2  φ_3  φ_4  φ_5  φ_6 

How to select reduced integration points for oscillatory basis functions?

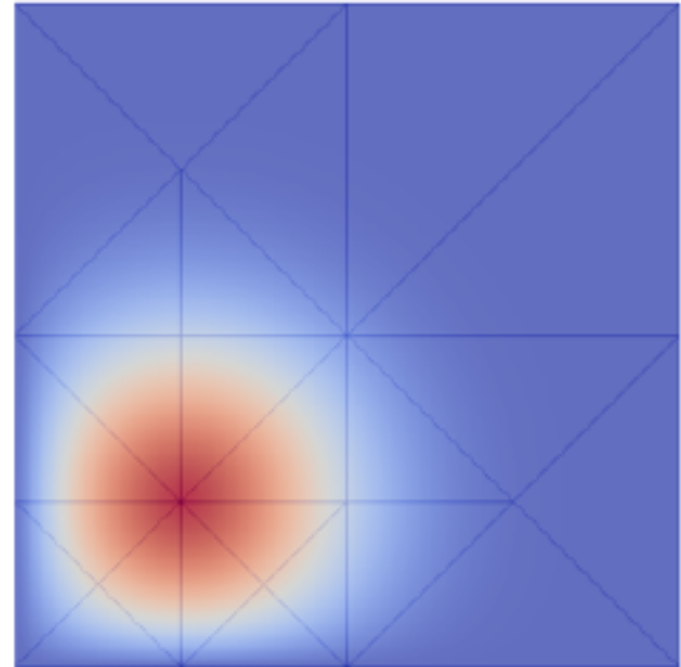
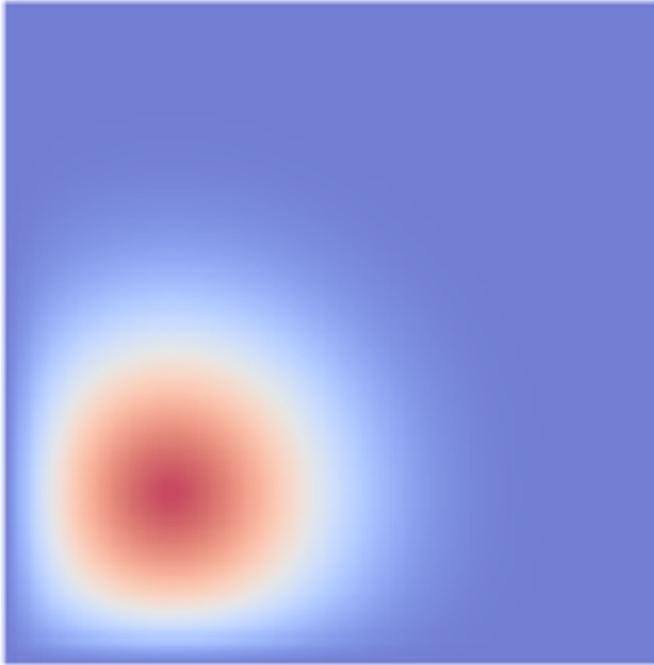


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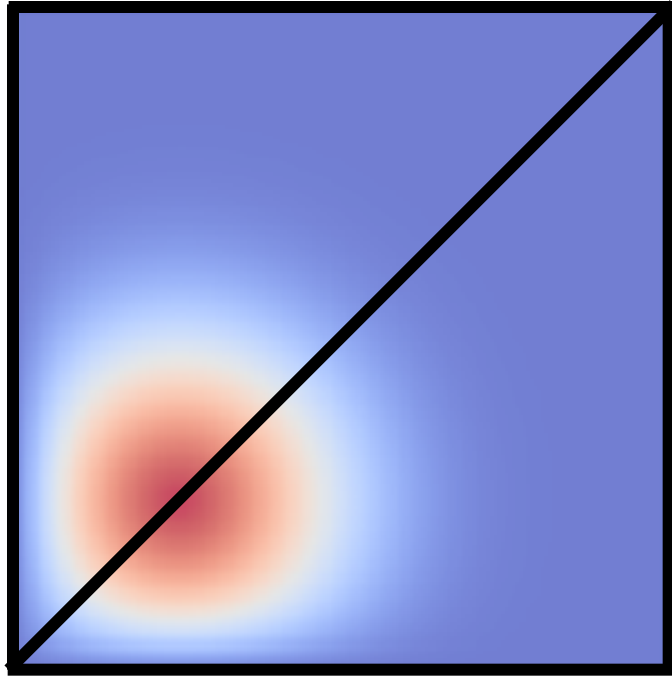
Locally approximate the modes linearly!



Let's look at φ_1

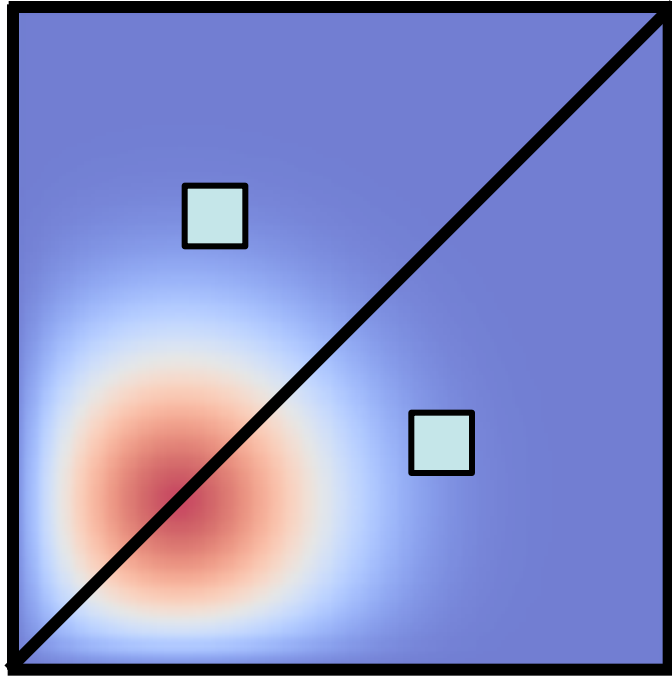


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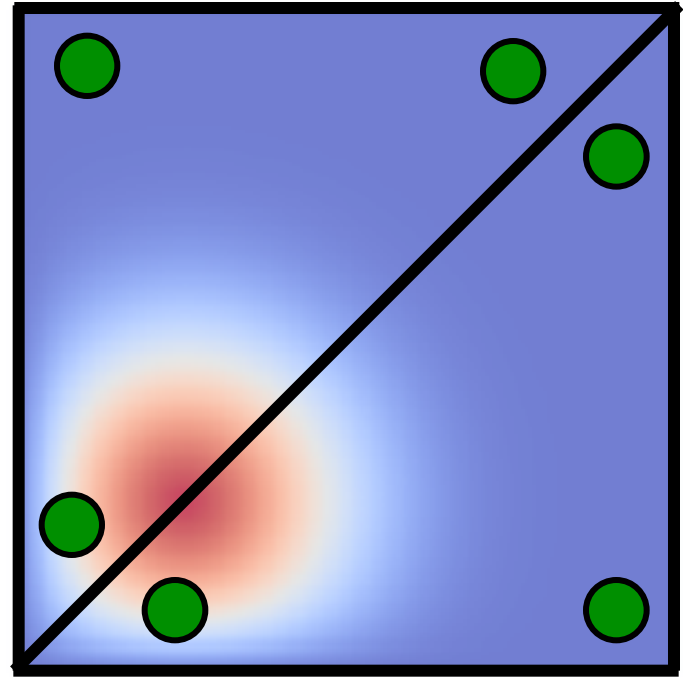
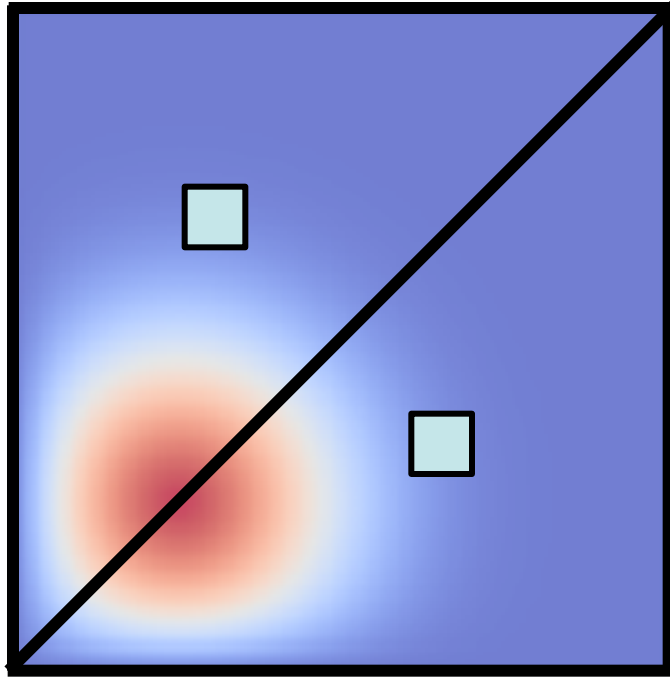


Start with coarse mesh

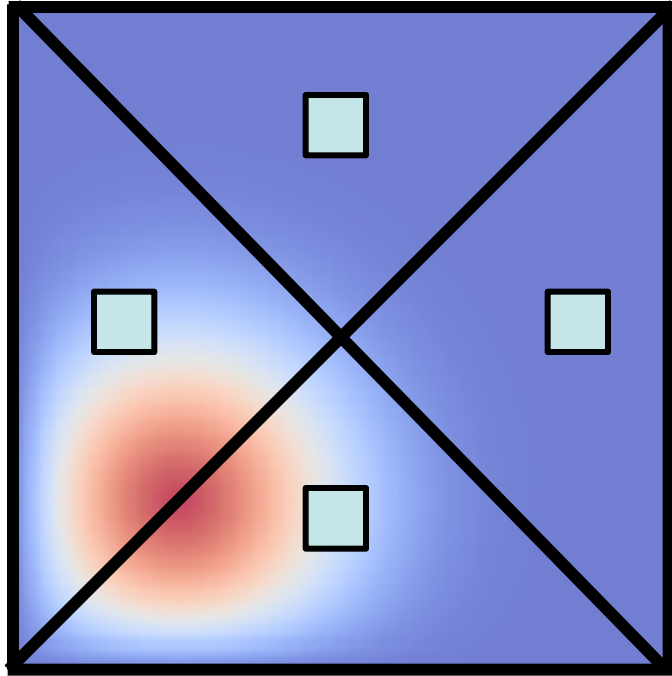
Let's look at φ_1



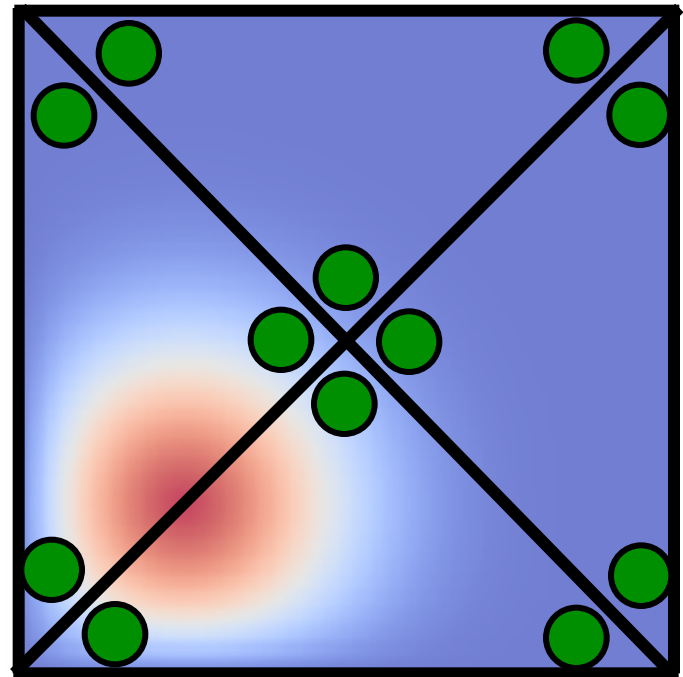
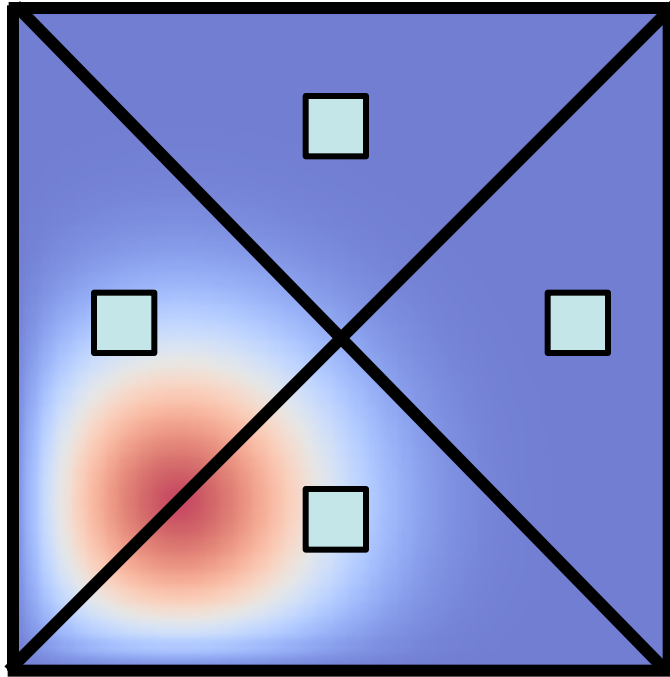
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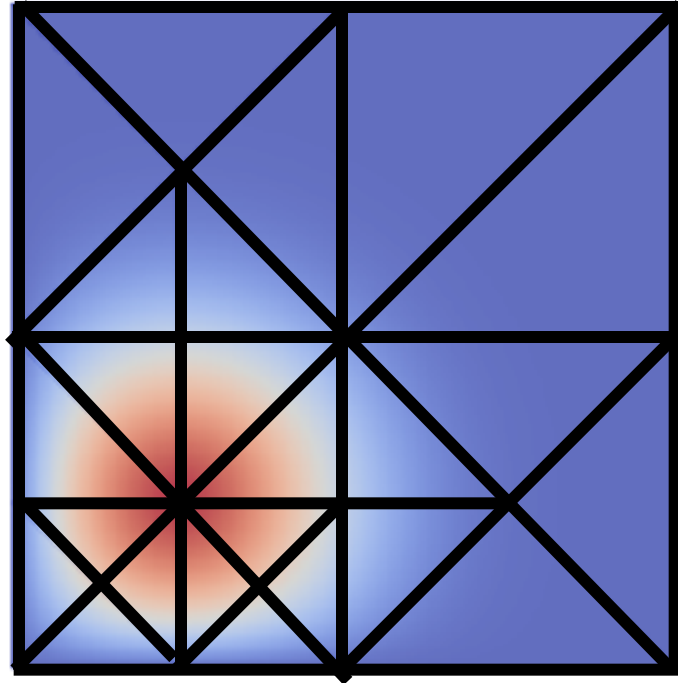


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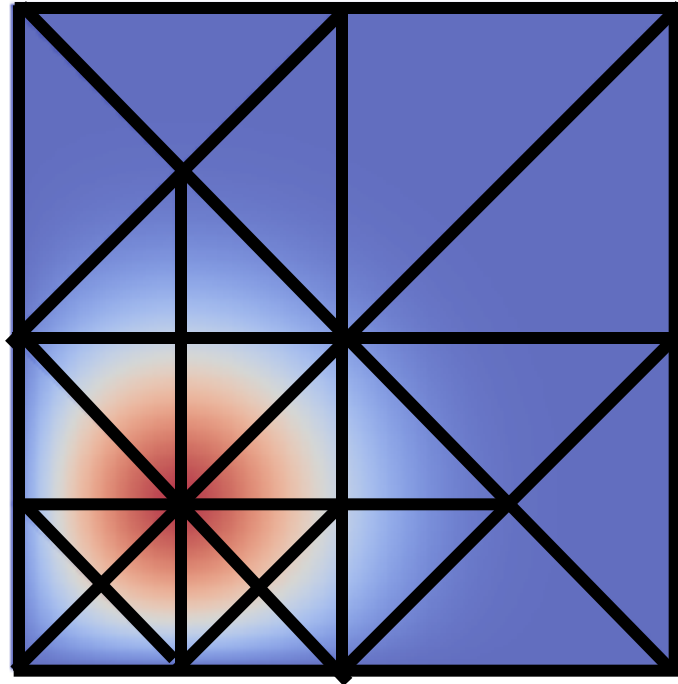


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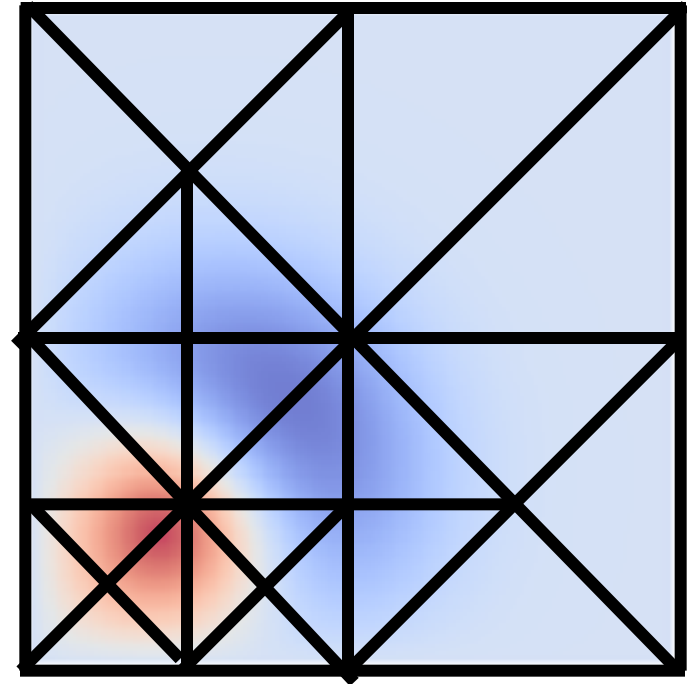




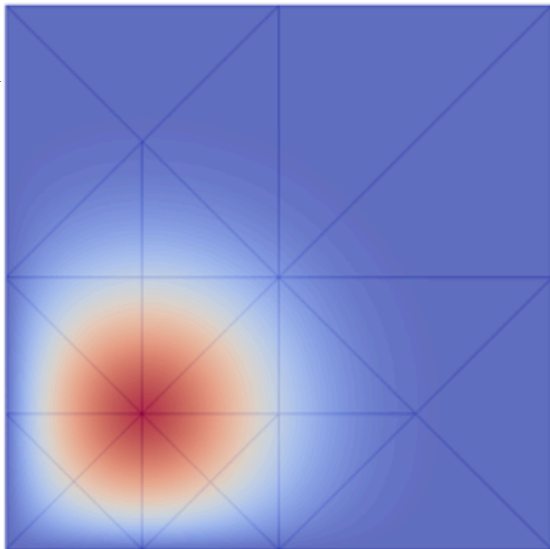
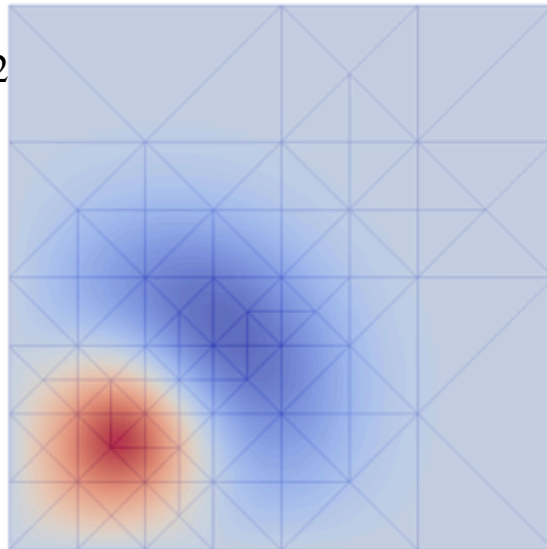
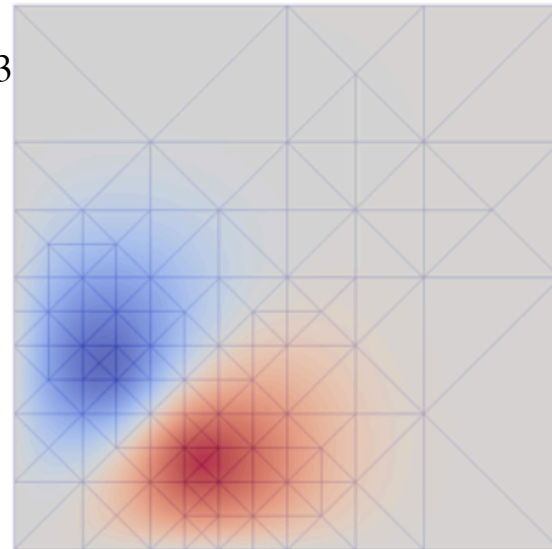
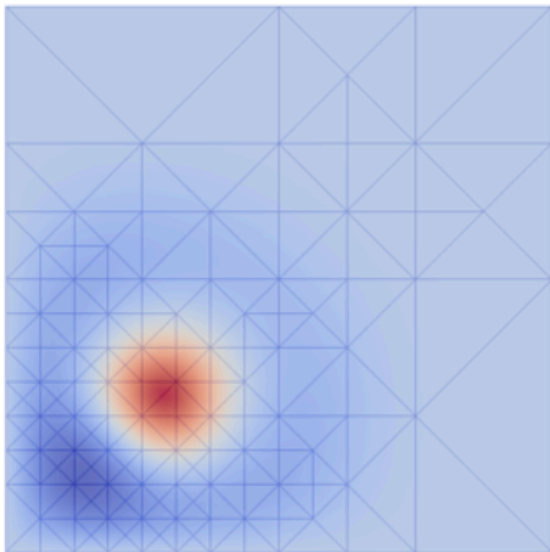
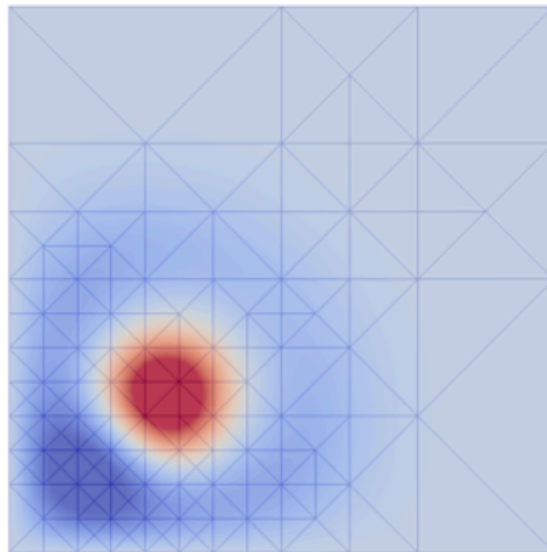
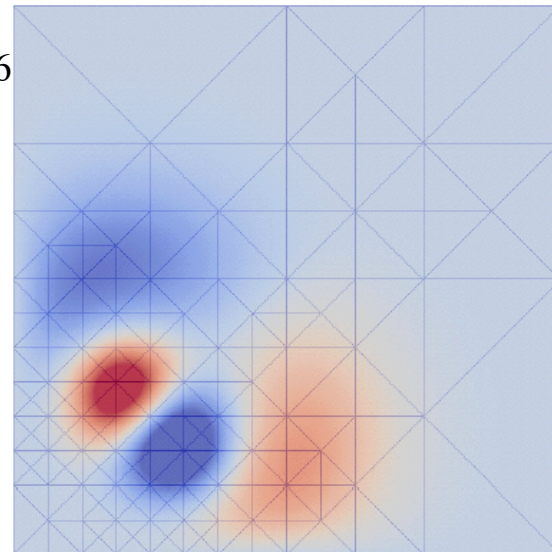
Final mesh of φ_1



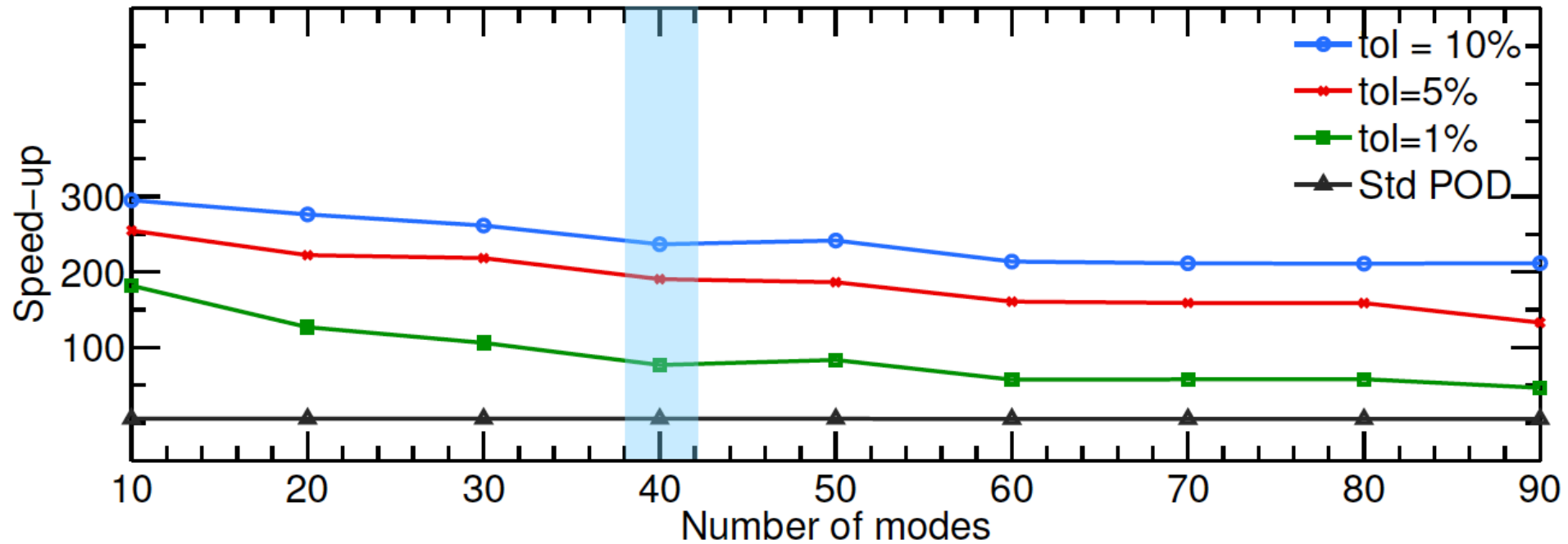
Final mesh of φ_1



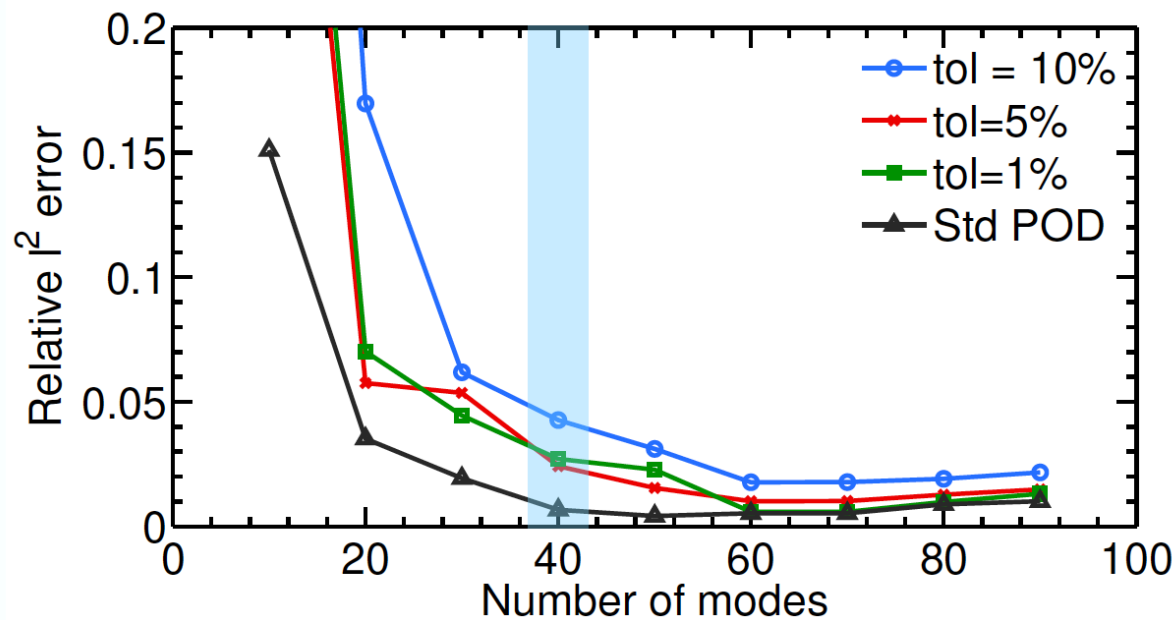
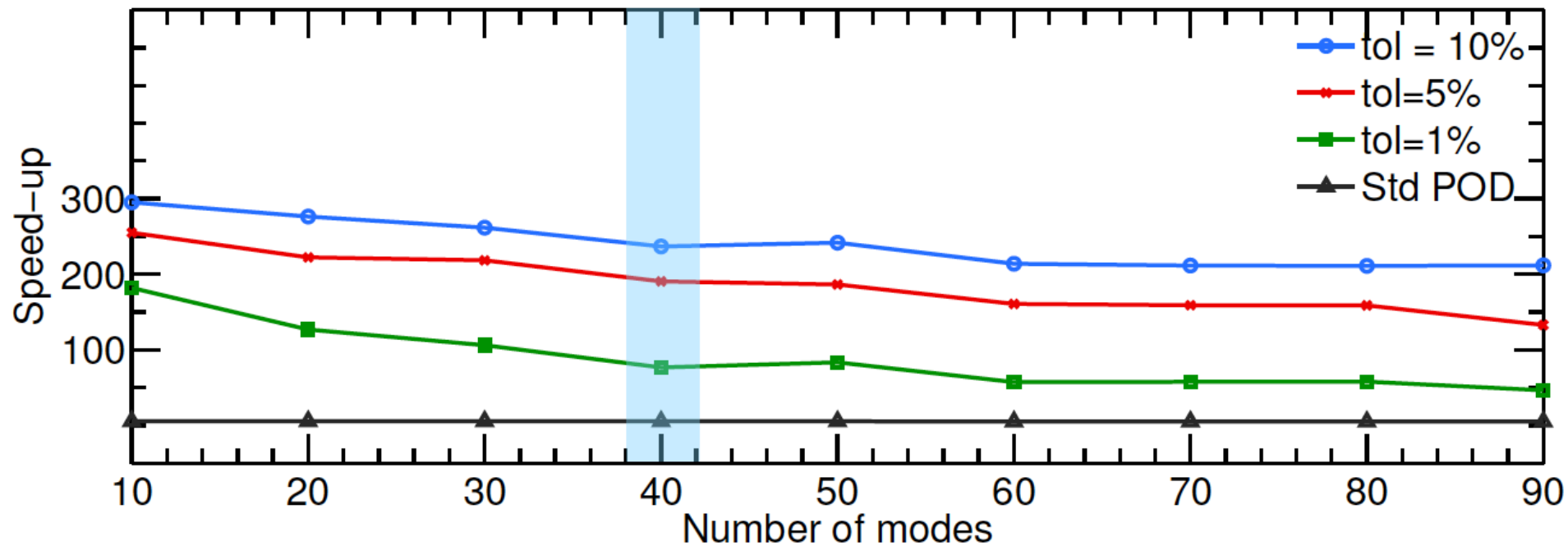
is the first mesh for φ_2

φ_1  φ_2  φ_3  φ_4  φ_5  φ_6 

POD-based ROM

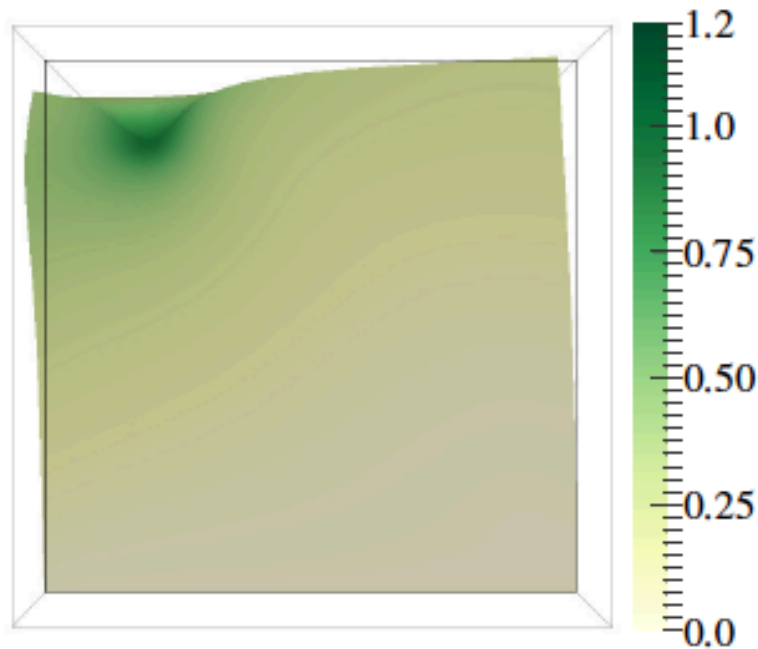


POD-based ROM

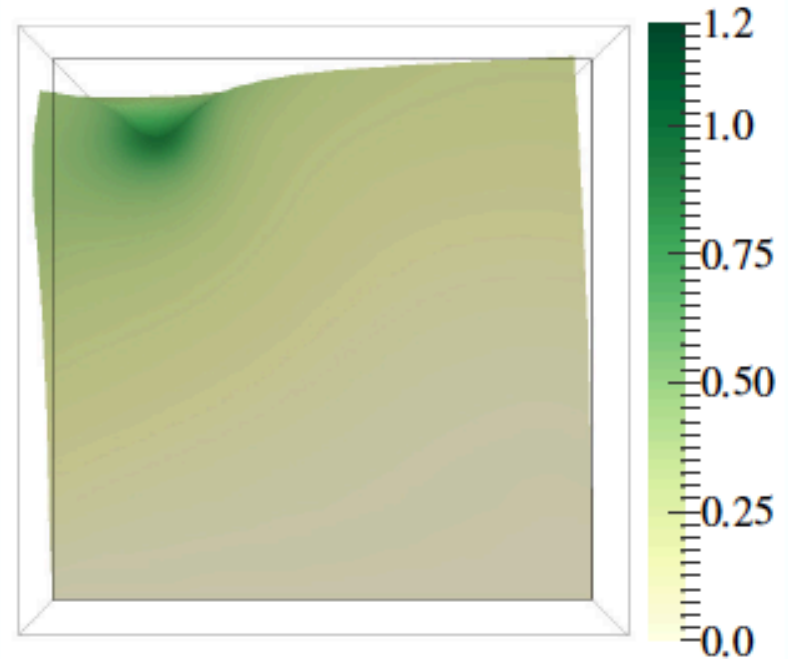


Indentation of a hyperelastic cube

30x faster than standard POD



FEM



Reduced integrated POD

Concluding remarks

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Type of discretisation	ONLY REGULAR	ANY

- 1. QC for irregular networks**
- 2. (goal-oriented) Adaptivity for QC**
- 3. Apply QC to true materials, no academic examples**
- 4. Geometrical scaling of POD-modes**
- 5. Coupling of POD domains**